

Statistical Graphics in Drug Development

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Insightful Life Sciences Event
April 7, 2006
Bethesda, Maryland

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Outline

1 Introduction

- Principles of Good Graphics
- Bad Graphic Examples

2 Use of Static Graphics

- Data Exploration
- Data Visualization in Presentation of Data
- Statistical Models

3 Static Graphics to Interactive

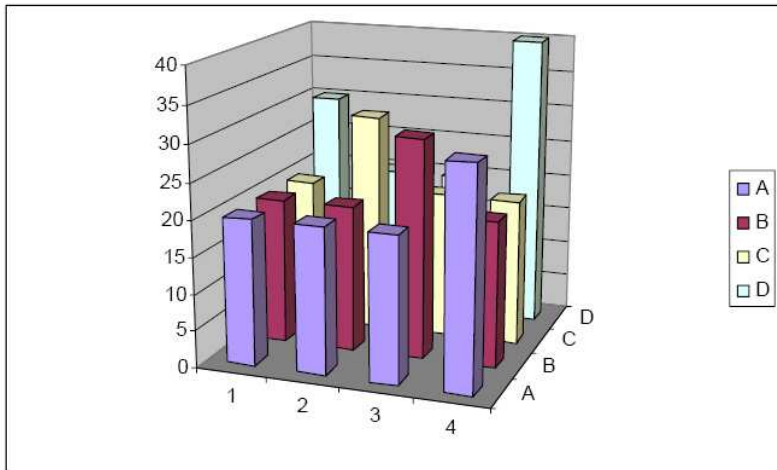
4 Conclusion

From the Minds of Cleveland and Tufte

- Exclude unneeded dimensions.
- Increase ink ratio $[(\text{data-ink})/(\text{total ink in graph})]$.
- Graphs do *not* have to depict important information quickly...may require detailed study.
- Omit “chartjunk”
- Keep continuous variables continuous.
- Show all the data when possible.

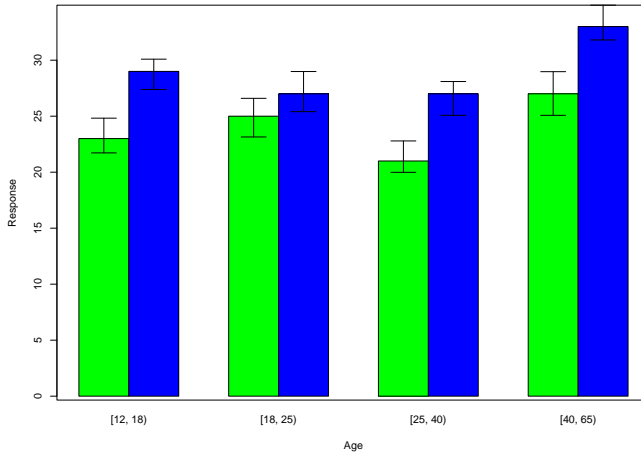
Bad Graphic Examples

Example One: 3-D Barplots



Bad Graphic Examples

Example Two: Barplot w/ Cont. Indep. Variable



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Introduction

Visual Exploration of the Data

- Patient Profiles
 - Useful for small studies.
- Look for any interesting features.
 - Outliers
 - Missing Data
 - Etc.
- Do results vary by baseline characteristics?
- Use all the data when possible.
 - Superpose summary statistics.
- Goal is to understand the data.

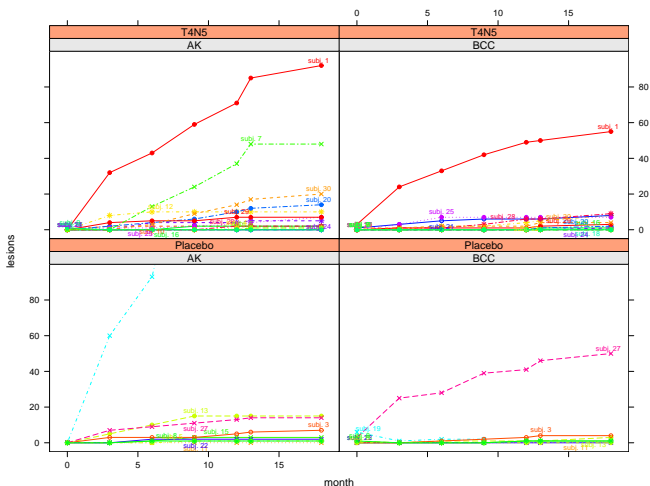
Patient Profile Example

Lancet Data for XP (Yarosh *et al.*, 2001)

- Study Objective: Does T4N5 prevent growth of new AK or BCC lesions?
- Two treatments: T4N5 and vehicle.
- Small sample sizes: T4N5=20, vehicle=9.
- Author's claim treatment effect.

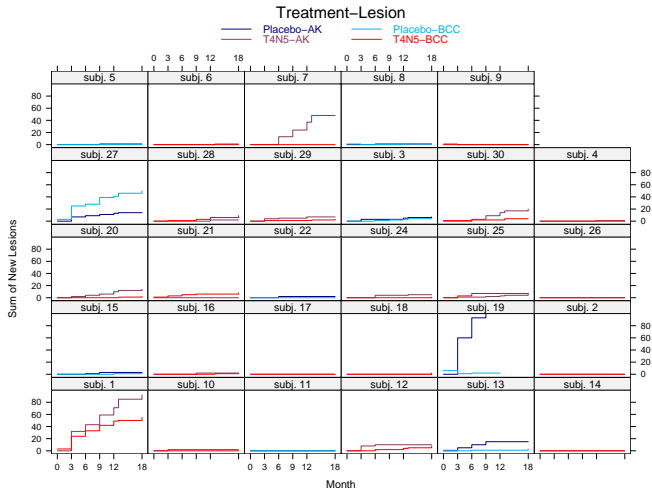
Data Exploration

Lancet Results for XP



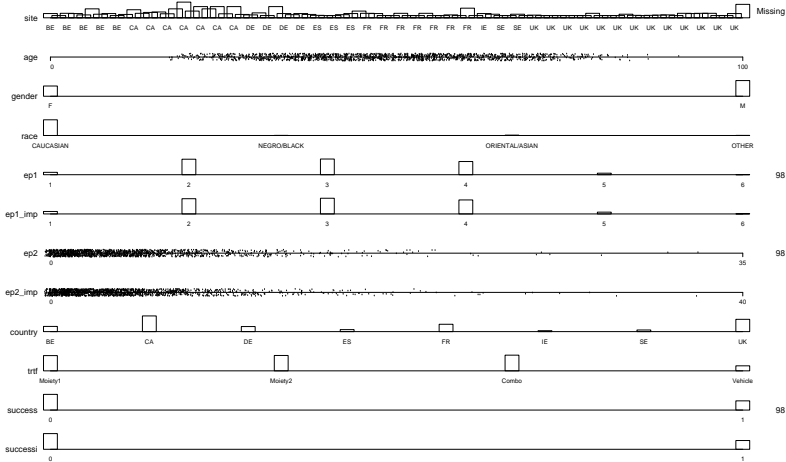
Data Exploration

Lancet Results for XP, cont.



Data Exploration

Data Density Plot



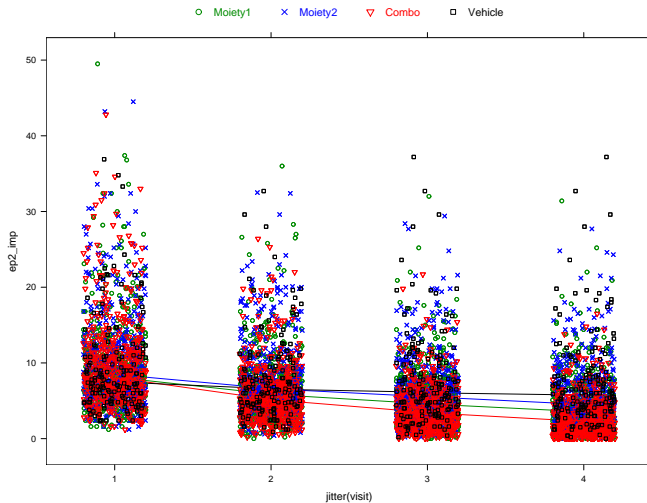
S Code

```
dat <- read.xport("C:/Teaching/Data/efficacy.xpt")  
names(dat) <- tolower(names(dat))
```

```
wk4 <- dat2[dat2$visit==4,]  
wk4d <- dat2[dat2$visit==4,-c(1,2,3,5,13,18)]
```

```
datadensity(wk4d) #Frank Harrell's Hmisc package
```

Efficacy Across Time



S Code

```
xyplot(ep2_imp ~ jitter(visit), groups=trtf, data=dat2,  
  panel=function(x,y,...){  
    panel.superpose(x,y,...)  
    panel.plsmo(x,y,...)  
  },  
key=list(  
  points=list(  
    pch=trellis.par.get("superpose.symbol")$pch[1:4],  
    col=trellis.par.get("superpose.symbol")$col[1:4]),  
  text=list(levels(dat2$trtf),  
    col=trellis.par.get("superpose.symbol")$col[1:4]),  
  columns=4))
```

Introduction

Objective: Use Data to Tell Story

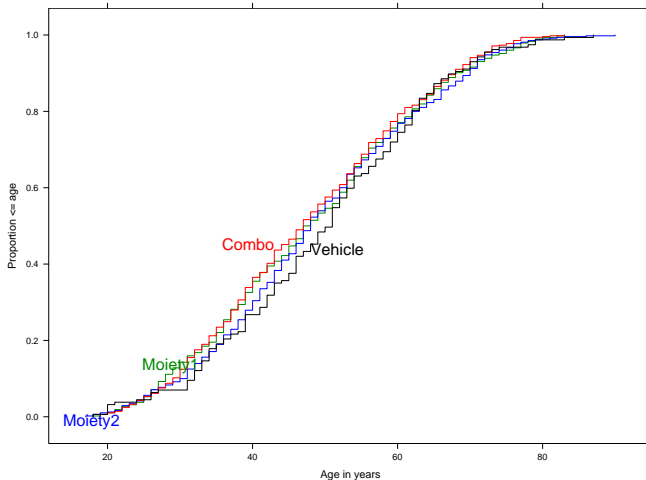
- Present and convey results.
 - Graphics and statistical results intertwined.
 - Graphics supplant large tables of information but not meant to replace simple tables.
- Use all data when possible.
 - Don't rely on means and SD's for continuous variables.
 - Display distribution of data if possible.
- Examine relationships between variables
 - Panel on factor variable(s)

Distribution of Age at Baseline—Not the Full Story

Typical Summary Table of Age at Baseline

| Treatment | Mean | SD |
|-----------|------|------|
| Combo | 47.6 | 14.4 |
| Moiety1 | 48.2 | 15.0 |
| Moiety2 | 48.9 | 14.7 |
| Vehicle | 49.8 | 14.4 |

Distribution of Age at Baseline–Full Story

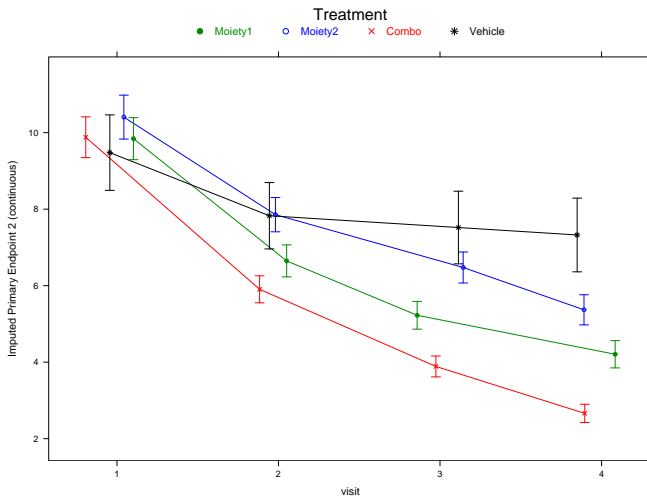


S Code

`summarize`, `smean.sd` and `ecdf` functions are available in Frank Harrell's `Hmisc` package.

```
base <- dat2[dat2$visit==1,]  
with(base, summarize(age, llist(trtf), smean.sd))  
  
ecdf(~age, groups=trtf, data=base)
```

Efficacy Across Time, Summary



S Code

```
ss <- with(dat2, summarize(ep2_imp, llist(visit,trtf),  
                             smean.cl.normal))  
names(ss) <- Cs(visit, trt, mean, lower, upper)  
  
xYplot(Cbind(mean,lower,upper)~jitter(visit),  
        groups=trt, data=ss)
```

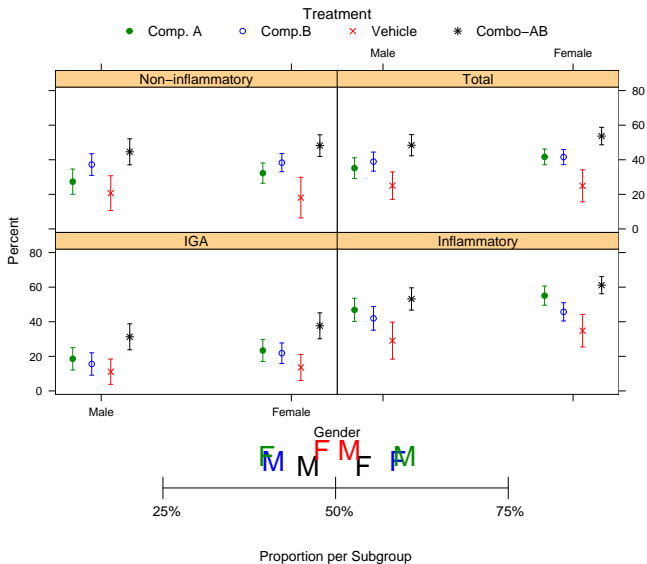
Examples Motivated by Dermatology Products

- 1 Efficacy results by subgroup for multiple endpoints.
 - Alternative to large tables.
- 2 HPA axis suppression by age cohort.
 - Assess damage to immune system function
- 3 Local Topical safety.
 - Assess change across time.
- 4 Time to First Event.
 - Alternative to basic AE rate tables.

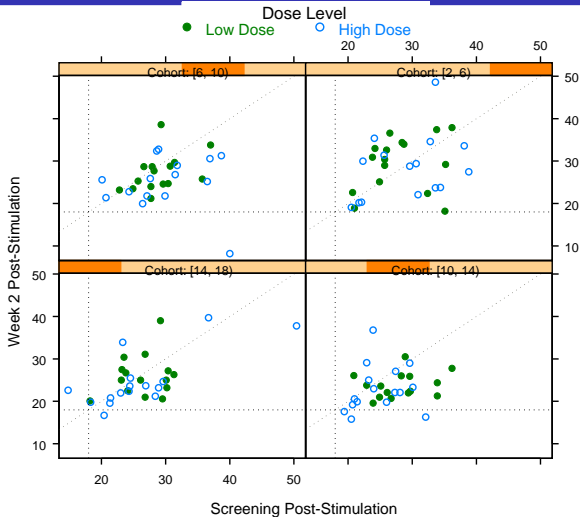
Efficacy by Subgroup Results

| | Combo (AB) | MoietyA | MoietyB | Vehicle |
|--|-------------|-------------|-------------|-------------|
| Total Lesions | | | | |
| Male | 46.5 (33.2) | 31.8 (32.2) | 39.0 (31.2) | 43.2 (30.2) |
| Female | 42.3 (28.7) | 27.8 (28.7) | 25.8 (28.5) | 24.1 (21.4) |
| Inflammatory Lesions | | | | |
| Male | 48.3 (43.8) | 34.6 (33.6) | 26.5 (34.2) | 19.7 (28.7) |
| Female | 51.1 (32.5) | 46.7 (28.7) | 41.7 (31.5) | 38.7 (32.5) |
| Non-inflammatory Lesions | | | | |
| Male | 46.5 (33.2) | 31.8 (32.2) | 39.0 (31.2) | 43.2 (30.2) |
| Female | 42.3 (28.7) | 27.8 (28.7) | 25.8 (28.5) | 24.1 (21.4) |
| Investigator's Global Assessment Success | | | | |
| Male | 48.3 (43.8) | 34.6 (33.6) | 26.5 (34.2) | 19.7 (28.7) |
| Female | 51.1 (32.5) | 46.7 (28.7) | 41.7 (31.5) | 38.7 (32.5) |

Data Visualization in Presentation of Data

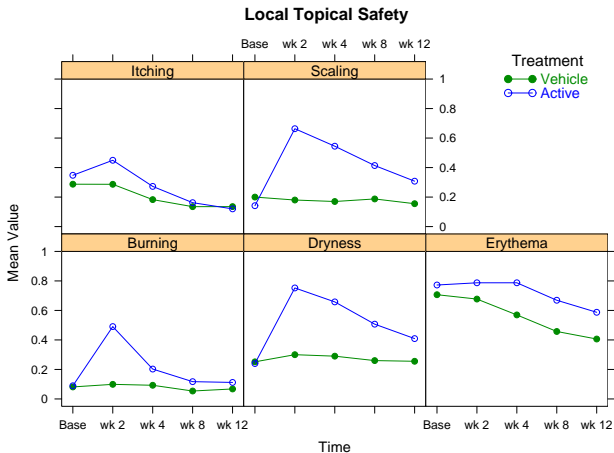


HPA Axis Suppression



** Points below horizontal line are suppressed

Local Safety

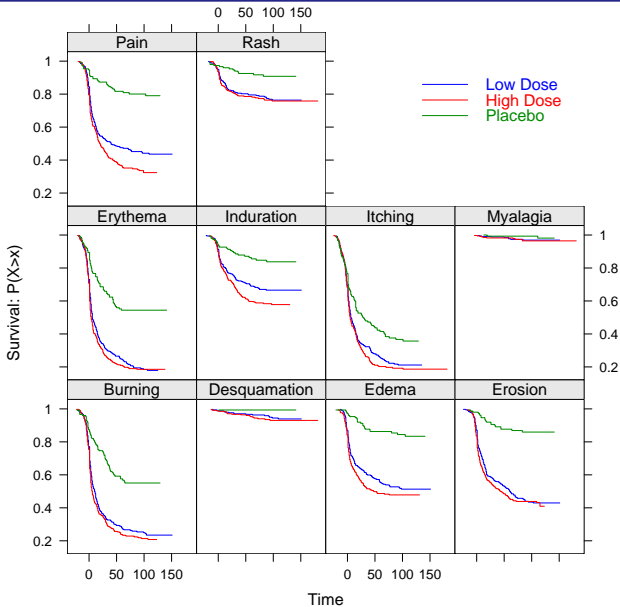


Time to First Event

Typical Summary Table of Adverse Events by Treatment

| AE | Low Dose | High Dose | Vehicle |
|----------|-----------|-----------|-----------|
| Pain | x_L (%) | x_H (%) | x_v (%) |
| Rash | x_L (%) | x_H (%) | x_v (%) |
| Erythema | x_L (%) | x_H (%) | x_v (%) |
| Burning | x_L (%) | x_H (%) | x_v (%) |
| \vdots | \vdots | \vdots | \vdots |

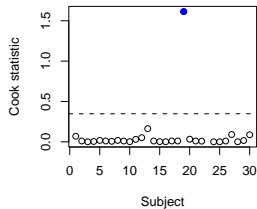
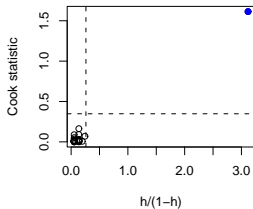
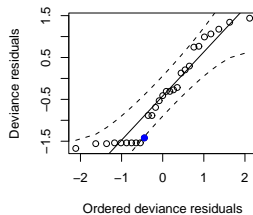
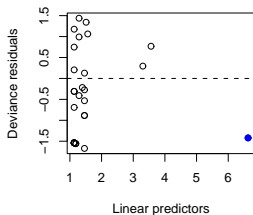
Data Visualization in Presentation of Data



Introduction

- Display model diagnostics.
 - QQ plots
 - Residual plots
 - etc.
- Depict model graphically.
 - Include confidence bands around the fit.
 - Can be useful for complex models.

Model Diagnostics



Comparing Two Psoriasis Endpoints

- Endpoint in US: Investigator Global Assessment (IGA)
 - Typically a 6 point scale: 0=clear to 5=severe
 - Each grade provides descriptions of skin parameter (e.g. scaliness, thickness, color)
- Endpoint in Europe: Psoriasis Area and Severity Index (PASI)
 - Takes into account:
 - Extent (E) involved (0-6)
 - Symptoms: redness (R), thickness (T), and scaliness (S); values 0=absent - 4=severe
 - Body Location: Arms = $.2 * (R + T + S) * E$,
Trunk = $.3 * (R + T + S) * E$, Legs = $.4 * (R + T + S) * E$
 - PASI is sum of all scores (values 0 - 64.8)
- One pivotal Phase 3 trial with IGA and PASI
- Five Phase 3 trials with PASI only.

Statistical Models

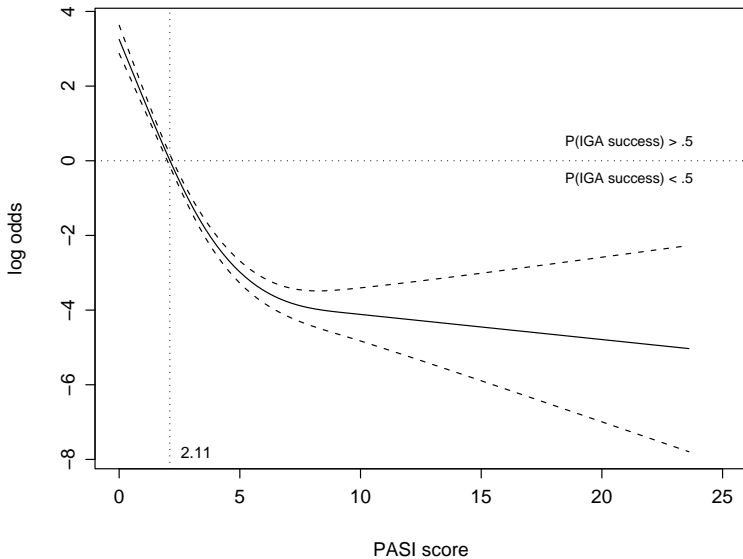
- Logistic regression model to predict IGA success (clear or almost clear) using PASI score at the end of treatment.
- Use a restricted cubic spline on PASI with 3 knots (lrm function in Design library)
- Estimated Model:

$$P(\text{IGA Success}) = \frac{1}{1 + \exp\{-X\hat{\beta}\}}, \quad \text{where}$$

$$\hat{\beta} = 3.25 - 1.58X + .02(X - 0.6)_+^3 - .03(X - 3.3)_+^3 + .01(X - 9.4)_+^3$$

$$\text{and } (x)_+ = \begin{cases} x & x > 0 \\ 0 & \text{OW} \end{cases}$$

Statistical Models



S code

Use Frank Harrell's Hmisc and Design libraries.

```
dd <- datadist(iga.win, pasi)
options(datadist='dd')

f <- lrm(iga.win ~ rcs(pasi, 3), data= reln, x=T, y=T)
anova(f) # Performs Wald tests

plot(f, xlab='Percent Change in PASI')
box()
abline(h=0, lty=3)
text(37, .5, labels="P(IGA success) > .5", cex=.75)
text(37, -.5, labels="P(IGA success) < .5", cex=.75)
```

Statistical Models

Now that we're cookin', as Emeril Lagasse would say,
'It's time to kick it up a notch!'



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Introduction

Add Capability of Interacting with the Graphic, Splus GraphletsTM

- Ability to show multivariate structures
 - Safety AND Efficacy together
- Access to multitudes of information
 - Include hyperlinks
- Multi-disciplinary tool
 - Data exploration at the fingertips of end-user (M.D.)
 - No software requirement to use other than web browser with Java capabilities.

DSI Consult

Objective: Identify any Influential Sites for Inspection

- Depict overall efficacy and efficacy within site
- Show sample size within site
- Be able to assess adverse events by System Organ Class (SOC) for each site
- Audience: Review Team (Medical, Stat, DSI)
- `E:/TEMP/soukup/DSIconult/Introduction.html`

HPA Axis Suppression

Objective: Assess Safety of Potent Topical Corticosteroid

- HPA: Method of assessing immune system function for potent topical corticosteroids.
 - Serum cortisol levels $\leq 18\mu\text{g}/\text{dL}$ at 30 minutes post-stimulation considered suppressed.
 - Because of a higher ratio of skin surface area to body mass, it is theorized that children are at a greater risk of suppression.
- Examine reported AE's for individual subjects.
- E: /TEMP/soukup/HPA/HPA.html

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- 1 Graphics are critical in conveying safety and efficacy information obtained in clinical trials.
- 2 No single plot is correct for all situations—let data guide you.
- 3 Use graphics throughout data assessment.
 - Exploration
 - Means of conveying information
 - Tool to interact
- 4 Intertwine graphics, tables, and statistics in reports.
 - Use good graphical principles.
 - Tables can be effective for small amounts of data.
- 5 Graphic creation is easily handled in the S language.

Acknowledgements

- Frank Harrell, Vanderbilt University
- Matt Austin, Amgen
- Michael O'Connell, Insightful
- Clinical Review Team in Dermatology and Dental Products (DDP), FDA
- Statistical Review Team supporting DDP, FDA

Sources

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